FACULTY OF
Humanities
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## Graphs

Data Structures and Algorithms for CL III, WS 2019-2020

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Data Structures \& Algorithms in Python



> 14.1 Graphs * The Graph ADT
> 14.2 Data Structures for Graphs
> * Edge List Structure
> * Adjacency List Structure
> * Adjacency Map Structure
> * Adjacency Matrix Structure

## Co-authorship Graph - undirected graph



[^0] Eigenvector centrality and represented by size.

Image from Alex Garnett, Grace Lee and Judy Illes. 2013. Publication trends in neuroimaging of minimally conscious states. PeerJ.
Search
OrthForm or Id: Katze $\quad \checkmark$ Ignore Case Find

## GermaNet Graph <br> - directed graph

From
http://www.sfs.uni-
tuebingen.de/Isd/documents/illustrations/ GernEdiT-screenshot-large.gif

Synsets

48836 nomen Tier [Katze]

50696 nomen Tier [Katze]


Lexical Units


$\square$

| Conceptual Relations Editor | Graph with Hyperonyms and Hyponyms | Lexical Relations Editor | Examples and Frames |
| :--- | :--- | :--- | :--- |

Hyperonyms and Hyponyms


## City Map - mixed graph




## DBpedia

## Mixed graph

http://dbpedia.org/page/Berlin
http://en.lodlive.it/?http\%3A\%2F\%2Fdbpedia.org\%2Fresource\%2FBerlin

## Graphs

- A graph $G$ is a set $V$ of vertices - together with a collection $E$ of pairwise connections between vertices from $V$, called edges
- Graphs are a way of representing relationships that exist between pairs of objects

- Edges in a graph are either directed or undirected
- An edge $(u, v)$ is directed from $u$ to $v$ if the pair $(u, v)$ is ordered, with $u$
 preceding $v$
- An edge $(u, v)$ is undirected if the pair $(u, v)$ is not ordered



## Types of Graphs

- undirected graph: all the edges in the graph are undirected

- directed graph (digraph): all the edges in the graph are directed
- mixed graph: has both directed and undirected edges


## Graph Terminology

- Two vertices joined by an edge are called the end vertices/endpoints of the edge
- $u$ and $v$ are the endpoints of edge 1
- Two vertices $u$ and $v$ are adjacent if there is an edge whose end vertices are $u$ and $v$
- $v$ and $x$ are adjacent
- An edge is called incident to a vertex if the vertex is one of the edge's endpoints
- edges 1, 2 and 4 are incident to $v$
- The degree of a vertex, $\operatorname{deg}(v)$, is the number of incident edges of $v$ : $v$ has degree 3
- Edges with the same endpoints are called parallel edges:

- 8 and 9 are parallel edges
- An edge is a self-loop is its two endpoints coincide:
- 10 is a self-loop


## Graph Terminology (cont'd)

- A path is a sequence of alternating edges and vertices that
- Starts with a vertex
- Ends with a vertex
- Each edge is incident to its predecessor and successor vertex
- A path is simple if each vertex in the path is distinct
- Examples of paths
- $P_{1}=(V, b, X, h, Z)$ is a simple path
- $P_{2}=(U, c, W, e, X, g, Y, f, W, d, V)$ not a simple path because $W$ appears twice



## Graph Terminology (cont'd)

- A cycle is a path that
- Starts and ends at the same vertex
- Includes at least one edge
- A cycle is simple if all its vertices are distinct, except for the first and the last vertex
- Examples of cycles
- $C_{1}=(V, b, X, g, Y, f, W, c, U, a, V)$ is a simple cycle
- $C_{2}=(U, c, W, e, X, g, Y, f, W, d, V, a, U)$ is not a simple cycle because $C_{2}$ goes twice through $W$



## Graph Terminology (cont'd)

- A vertex $u$ reaches a vertex $v$, and $v$ is reachable from $u$ if there is a path from $u$ to $v$
- $u$ reaches $y$ in $G_{1}$
- $u$ does not reach $b$ in $G$
- A graph is connected if for any two vertices there is a path between them

- $G_{1}$ and $G_{2}$ are connected graphs
- $G$ is not a connected graph
- A subgraph of a graph of $G$ is a graph whose vertices and edges are subsets of the vertices and edges of $G$
- $G_{1}$ and $G_{2}$ are subgraphs of $G$
- If a graph is not connected, its maximal connected subgraphs are called the connected components of $G$
- $G_{1}$ and $G_{2}$ are the connected components of $G$


## Graph Terminology (cont'd)

- a spanning subgraph of a graph $G$ is a subgraph of $G$ containing all the vertices of $G$
- A forest is a disconnected graph without cycles
- A tree is a connected forest - that is - a connected graph without cycles
- A spanning tree of a graph is a spanning subgraph that is a tree

spanning subgraph

spanning tree

(h)



## Graph Properties

- Property 1 . If $G$ is a graph with $m$ edges and vertex set $V$, then

$$
\sum_{v \in V} \operatorname{deg}(v)=2 m
$$

- Justification. Any edge $(u, v)$ is counted twice in the summation:
- Once for its endpoint $u$
- Once for its endpoint $v$
- The total contribution of the edges to the degrees of the vertices is twice the number of edges.


## Graph Properties (cont'd)

- Property 2. If $G$ is a simple undirected graph with $n$ vertices and $m$ edges, then

$$
m \leq \frac{n(n-1)}{2}
$$

- Justification. $G$ is simple, meaning that -
- there are no edges that have the same endpoints (no parallel edges)
- there are no self-loops
- then the maximum degree of a vertex in $G$ is $n-1$
- according to property $1,2 m \leq n(n-1) \Rightarrow m \leq \frac{n(n-1)}{2}$


## The Graph ADT

## The Graph ADT

- A graph is a collection of vertices and edges
- Can be modelled as a combination of three data types: Vertex, Edge and Graph
- class Vertex
- Lightweight object storing the information provided by the user
- The element() method provides a way to retrieve the stored information
- class Edge
- Another lightweight object storing an associated object - the cost
- The element() method provides a way to retrieve the cost of the edge
- endpoints() method: returns a tuple ( $u, v$ ) where $u$ and $v$ are the Vertex objects
- opposite( $v$ ) method: assuming vertex $v$ is one endpoint of an edge, return the other endpoint


## The Graph ADT (cont'd)

- class Graph: can be either undirected or directed - flag provided to the constuctor

| vertex_count() | returns the number of vertices of the graph |
| :--- | :--- |
| vertices() | returns an iteration of all the vertices of the graph |
| edge_count() | returns the number of edges of the graph |
| edges() | returns an interation of all the edges of the graph |
| get_edge(u,v) | returns the edge from vertex $u$ to vertex $v$, if one exists, otherwise None |
| degree(v) | returns the number of edges incident to vertex $v$ |
| incident_edges(v) | returns an iteration of all edges incident to vertex $v$ |
| insert_vertex(v, x=None) | create and return a new Vertex storing element $x$ |
| insert_edge(u,v,x=None) | create and return a new Edge from vertex $u$ to vertex $v$, storing $x$ |
| remove_vertex(v) | remove vertex $v$ and all its incident edges from the graph |
| remove_edge(e) | remove edge $e$ from the graph |

## Data Structures for Graphs

## Data Structures for Graphs

- Four data structures for representing a graph

1. Edge list
2. Adjacency list
3. Adjacency map
4. Adjacency matrix

- In each representation
- Same: maintain a collection to store the vertices of a graph
- Different: organize the edges


## Edge List Structure

- In an edge list, we maintain
- an unordered list $V$ to store all vertex objects
- an unordered list $E$ to store all edge objects
- To support the methods of the Graph ADT, assume:

- Vertex
- A reference to element $x$ to support the element() method
- A reference to the position of the vertex instance in the list $V$ - for efficient vertex removal
- Edge
- A reference to element $x$, to support the element() method
- A reference to the position of the edge instance in list $E$ - for efficient edge removal
- References to the vertex objects associated with the endpoints of $e$



## Edge List Structure (cont'd)

- In an edge list, we maintain
- an unordered list $V$ to store all vertex objects
- an unordered list $E$ to store all edge objects
- A very simple structure, though not very efficient:

- locating a particular edge ( $u, v$ ) - traversing the entire edge list
- obtaining the set of all edges incident to a vertex $v$ - again, traverse then entire edge list



## Edge List Structure - Performance

- Space usage
- $O(n+m)$ for a graph with $n$ vertices and $m$ edges
- Assuming each individual vertex or edge uses $O$ (1) space
- The lists $V$ and $E$ use space proportional to their number of entries


## Edge List Structure - Performance (cont'd)

| Operation | Running Time |
| :--- | :--- |
| vertex_count( ), edge_count( ) | $O(1)$ |
| vertices() | $O(n)$ |
| edges( ) | $O(m)$ |
| get_edge(u,v), degree(v), incident_edges(v) | $O(m)$ |
| insert_vertex(x), insert_edge(u,v,x), remove_edge(e) | $O(1)$ |
| remove_vertex(v) | $O(m)$ |

- get_edge(u, v), degree(v), incident_edges(v) could be implemented more efficiently than $O(m)$
- remove_vertex (v) also entails removing all the edges incident to $v$ - otherwise the edges would point to a non-existing vertex of the graph - hence $O(\mathrm{~m})$


## Adjacency List Structure

- In an adjacency list, we maintain
- For each vertex, a separate list containing those edges tha are incident to the vertex
- To support the methods of the Graph ADT, assume:

- Vertex
- A reference to element $x$ to support the element() method
- A reference to the position of the vertex instance in the list $V-$ for efficient vertex removal
- A list $I(v)$ - the incidence list of $v$ - containing the edges that are incident to $v$
- Edge
- A reference to element $x$, to support the element() method
- References to the vertex objects associated with $e$ 's endpoints
- References to the positions of the edge instance in lists $I(u)$
 and $I(v)$ - for efficient edge removal


## Adjacency List Structure (cont'd)

- In an adjacency list, we maintain
- For each vertex, a separate list containing those edges tha are incident to the vertex
- Benefits compared to the edge list

- The $I(v)$ list of each node $v$ contains exactly the edges that should be reported by incident_edges(v)
- Iterate $I(v)$ in $O(\operatorname{deg}(v))$ time instead of iterating the full edge list - the best possible outcome for any graph representation, since there are $\operatorname{deg}(v)$ edges to report



## Adjacency List Structure - Performance

- Space usage: asymptotically, the same as the edge list structure
- $O(n+m)$ for a graph with $n$ vertices and $m$ edges
- The primary vertex list uses $O(n)$ space
- The sum of all secondary lists containing the edges incident to each vertex is $O(\mathrm{~m})$
- An undirected edge $(u, v)$ is referenced both in $I(u)$ and in $I(v)$, but its presence in the graph results only in a constant amount of additional space


## Adjacency List Structure - Performance (cont'd)

| Operation | Running Time |
| :--- | :--- |
| vertex_count( ), edge_count( ) | $O(1)$ |
| vertices( ) | $O(n)$ |
| edges( ) | $O(m)$ |
| get_edge(u,v) | $O(\min (\operatorname{deg}(u), \operatorname{deg}(v)))$ |
| degree(v) | $O(1)$ |
| incident_edges(v) | $O(\operatorname{deg}(v))$ |
| insert_vertex(x), insert_edge $(\mathrm{u}, \mathrm{v}, \mathrm{x})$ | $O(1)$ |
| remove_edge(e) | $O(1)$ |
| remove_vertex(v) | $O(\operatorname{deg}(v))$ |

- get_edge $(u, v)$ - we can look for the edge in either the list of $u$ or that of $v$ - take the shortest
- Because we are storing the positions of $e$ in $I(u)$ and $I(v)$, removing an edge takes $O(1)$ time
- To remove a vertex $v$ we need to also remove all its incident edges - but there are all in $I(v)$, so remove_vertex (v) runs in $O(\operatorname{deg}(v))$ time


## Adjacency Map Structure

- In an adjacency map, we maintain
- For each vertex $v$, a separate hash-map
- Each entry has as key the vertex that is opposite to $v$, and as value the edge which has $u$ and $v$ as endpoints
- To support the methods of the Graph ADT, assume:
- Vertex
- A reference to element $x$ to support the element() method
- A reference to the position of the vertex instance in the list $V$ - for efficient vertex removal
- A hashmap $I(v)$ - containing (vertex, edge) pairs where the vertices are the opposites of $v$ and the edges are the edges incident to $v$
- Edge
- A reference to element $x$, to support the element() method
- References to the vertex objects associated with $e$ 's endpoints



## Adjacency Map Structure

- In an adjacency map, we maintain
- For each vertex $v$, a separate hash-map
- Each entry has as key the vertex that is opposite to $v$, and as value the edge which has $u$ and $v$ as endpoints
- Benefits compared to the adjacency list
- get_edge(u,v) can be implemented in expected $O$ (1) time by searching for vertex $u$ as a key in $I(v)$ or vice-versa
- this is better than in the adjacency list case, where the best case performance was $O(\min (\operatorname{deg}(u), \operatorname{deg}(v)))$



## Adjacency Map Structure - Performance

- Space usage
- $O(n+m)$, just like the adjacency list
- For each vertex u, $I(u)$ - an adjacency map uses $O(\operatorname{deg}(u))$ space


## Adjacency Map Structure - Performance (cont'd)

| Operation | Edge List | Adj. List | Adj. Map |
| :--- | :--- | :--- | :--- |
| vertex_count () | $O(1)$ | $O(1)$ | $O(1)$ |
| edge_count () | $O(1)$ | $O(1)$ | $O(1)$ |
| vertices( ) | $O(n)$ | $O(n)$ | $O(n)$ |
| edges() | $O(m)$ | $O(m)$ | $O(m)$ |
| get_edge(u,v) | $O(m)$ | $O\left(\min \left(d_{u}, d_{v}\right)\right)$ | $O(1)$ exp. |
| degree(v) | $O(m)$ | $O(1)$ | $O(1)$ |
| incident_edges(v) | $O(m)$ | $O\left(d_{v}\right)$ | $O\left(d_{v}\right)$ |
| insert_vertex(x) | $O(1)$ | $O(1)$ | $O(1)$ |
| remove_vertex(v) | $O(m)$ | $O\left(d_{v}\right)$ | $O\left(d_{v}\right)$ |
| insert_edge(u,v,x) | $O(1)$ | $O(1)$ | $O(1)$ exp. |
| remove_edge(e) | $O(1)$ | $O(1)$ | $O(1) \exp$. |

- $d_{v}$ - the degree of $v$
- an adjacency map achieves essentially optimal running times for all methods, making in an excellent all-purpose choice as a graph representation structure


## Adjacency Matrix Structure

- In an adjacency matrix structure, we maintain
- An $n \times n$ matrix $A$ of edges, storing references to edges
- $A[i, j]$ stores a reference to the edge $(u, v)$ if it exists, where $u$ is the vertex with index $i$ and $v$ is the vertex with index $j$
- if there is no such edge, then $A[i, j]=$ None
- $A$ is symmetric if the graph is undirected
- An edge between a given pair of vertices can be retrieved in worst-case constant time



## Adjacency Matrix Structure - Performance

- Space usage
- $O\left(n^{2}\right)$ space, much worse than the $O(n+m)$ needed for the other three structures
- Although if the graph is dense the number of edges is proportional to $O\left(n^{2}\right)$
- In practice, most real-word graphs are sparse - making the adjacency matrix structure inefficient, since it will store many None values
- If a graph is dense, a adjacency matrix might be more efficient then an adjacency list or map
- Particularly if edges have no auxiliary data, then an adjacency matrix can be implemented using a Boolean matrix, using 1 bit to store information about each edge slot, e.g. $A[i, j]=$ True if and only if $(u, v)$ is an edge in the graph


## Adjacency Matrix Structure - Performance (cont'd)

| Operation | Edge List | Adj. List | Adj. Map | Adj. Matrix |
| :--- | :--- | :--- | :--- | :--- |
| vertex_count () | $O(1)$ | $O(1)$ | $O(1)$ | $O(1)$ |
| edge_count () | $O(1)$ | $O(1)$ | $O(1)$ | $O(1)$ |
| vertices () | $O(n)$ | $O(n)$ | $O(n)$ | $O(n)$ |
| edges () | $O(m)$ | $O(m)$ | $O(m)$ | $O(m)$ |
| get_edge(u,v) | $O(m)$ | $O\left(\min \left(d_{u}, d_{v}\right)\right)$ | $O(1)$ exp. | $O(1)$ |
| degree(v) | $O(m)$ | $O(1)$ | $O(1)$ | $O(n)$ |
| incident_edges $(\mathrm{v})$ | $O(m)$ | $O\left(d_{v}\right)$ | $O\left(d_{v}\right)$ | $O(n)$ |
| insert_vertex(x) | $O(1)$ | $O(1)$ | $O(1)$ | $O\left(n^{2}\right)$ |
| remove_vertex $(\mathrm{v})$ | $O(m)$ | $O\left(d_{v}\right)$ | $O\left(d_{v}\right)$ | $O\left(n^{2}\right)$ |
| insert_edge(u,v,x) | $O(1)$ | $O(1)$ | $O(1) \exp$. | $O(1)$ |
| remove_edge(e) | $O(1)$ | $O(1)$ | $O(1) \exp$. | $O(1)$ |

- get_edge(u,v) is an $O(1)$ operation
- Several operations are less efficient:
- degree(v), incident_edges(v) - we need to examine all $n$ entries in the row associated with vertex $v-O(n)$
- insert_vertex(v), remove_vertex(v) - the matrix has to be resized - $O\left(n^{2}\right)$


## Python Implementation - using an Adjacency Map variant

- Use a Python dictionary to represent each secondary incidence map, $I(v)$
- Use a top-level dictionary $D$ to map each vertex $v$ to its incidence map, $I(v)$
- All the vertices of the graph can be obtained by iterating over the keys of $D$
- This frees us from having to keep indices for the position of the vertices in the Vertex
- Also, rather than maintaining a separate list of edges, the edges can be found in $O(n+$ $m$ ) time by taking the union of the edges found in all the incidence maps


## Vertex class

```
#
class Vertex:
    """Lightweight vertex structure for a graph."""
    __slots__ = '_element'
    def __ init __(self, x):
        """Do not call constructor directly. Use Graph's insert_vertex(x)."""
        self._element = x
    @property
    def element(self):
        """Return element associated with this vertex."""
        return self._element
    def __hash__(self): # will allow vertex to be a map/set key
        return hash(id(self))
```


## slots

- By default Python represents each namespace with an instance dictionary of the built-in dict class- this maps identifying names in the scope to the associated objects
- While a dictionary structure supports relatively efficient name lookups, it requires additional memory beyond the raw data that it stores.
- Python provides a more direct mechanism for representing instance namespaces, that avoids the use of an auxiliary dictionary.
- To streamline the representation for all instances of a class, the class should define a class-level member named __slots__ that is assigned a fixed sequence of strings that serve as names for instance variables
- Advisable in particular in any nested classes that are expected to have many instances
.


## init

- Whenever an instance of the Vertex class is created using a statement of the type $v=$ Vertex("A"), a special method called $\qquad$ init $\qquad$ is called
- __init__ serves as the constructor of the class
- It is responsible primarily for establishing the state of the new object - e.g. set up the _element instance variable in the case of Vertex, set up the _origin, _destination and _element in the case of Edge
- By convention a single leading underscore in the name of a data member, such as _element implies that it is intended as nonpublic; users of a class should not directly access such members


## @property

- @property is a decorator which indicates that the element(self) method is a "getter" method, and that the name of the property is the method name only - e.g. only element
- A decorator is a function which receives another function as an argument
- The behavior of the argument function is extended by the decorator without actually modifying it
- The element of a vertex can then be obtained using x.element
- There is also a corresponding way of creating a setter using the @f.setter decorator

```
@element.setter
def element(self, el):
    self._element = el
```


## hash

- Standard Python mechanism for computing hash codes - hash(x) returns an integer value that serves as a hash code for object $x$
- Only immutable data types are hashable in Python - to ensure that the object's hash code remains constant during the lifetime of the object
- It an object is inserted into a hash table, and then its hash code would change, then a different object would be retrieved from the hash table
- Instances of user-defined classes are unhashable by default
- A function that computes the hash code can be implemented via the $\qquad$ hash $\qquad$ method within the class
- The returned hash code should reflect the immutable attributes of an instance (e.g. _element would not make for a good attribute for hashing, it might be updated)
- Also, if $x==y$, then hash $(x)==$ hash( $y$ )
\#nested Edge class
class Edge:
"""Lightweight edge structure for a graph."""
__slots_- = '_origin', '_destination', '_element'
def __init __(self, $u, v, x)$ :
""" Do not call constructor directly. Use Graph's insert_edge(u,v,x).""
self._origin $=u$
self..destination $=v$
self._element $=\mathrm{x}$
def endpoints(self):
"""Return ( $u, v$ ) tuple for vertices $u$ and $v . " "$ "
return (self._origin, self._destination)
def opposite(self, $\mathbf{v}$ ):
"""Return the vertex that is opposite von this edge."""
return self._destination if v is self._origin else self._origin
def element(self):
"""Return element associated with this edge.""
return self._element
def __hash __(self): \# will allow edge to be a map/set key
return hash( (self._origin, self._destination) )


## self

- self identifies the instance upon which a method is invoked
- self is also used to store the instance variables that reflect its current state
- self._element refers to an instance variable named _element that is stored as part of that particular Vertex's state
- There is a difference between a method signature as declared within a class vs. that used by a caller:
- E.g. from the user's perspective the opposite() method takes one parameter, the Vertex v, while endpoints() takes no parameters
- However, within the class definition self in an explicit parameter, making opposite() have two parameters, and endpoints() one parameter
- The Python interpreter will automatically bind the instance upon which the method is invoked to the self parameter

```
class Graph:
    ""Representation of a simple graph using an adjacency map.""
def __init__(self, directed=False):
    """Create an empty graph (undirected, by default).
    Graph is directed if optional paramter is set to True.
    self._outgoing = { }
    # only create second map for directed graph; use alias for undirected
    self..incoming = { } if directed else self._outgoing
    def is_directed(self):
    """Return True if this is a directed graph; False if undirected.
    Property is based on the original declaration of the graph, not its contents.
    return self._incoming is not self._outgoing # directed if maps are distinct
    def vertex_count(self):
    """Return the number of vertices in the graph."""
    return len(self._outgoing)
def vertices(self):
    ""Return an iteration of all vertices of the graph."""
    return self._outgoing.keys()
def edge_count(self):
    """Return the number of edges in the graph."""
    total = sum(len(self._outgoing[v]) for v in self._outgoing)
    # for undirected graphs, make sure not to double-count edges
    return total if self.is_directed( ) else total // 2
    def edges(self):
    """Return a set of all edges of the graph."""
    result = set( ) # avoid double-reporting edges of undirected graph
    for secondary_map in self._outgoing.values():
        result.update(secondary_map.values()) # add edges to resulting set
    return result
```


## Python Generators

- The most convenient technique for creating iterators in Python is through the use of generators
- A generator is implemented with a syntax that is very similar to a function, but instead of returning values, a yield statement is executed to indicate each element of a sequence
- It is illegal to combine return and yield statements in the same implementation
- Lazy evaluation: the results are only computed if requested, the entire sequence need not reside in memory at one time - generators can produce infinite sequences of values
- Generator comprehensions do not create temporary lists

Graph Class,
def get_edge(self, $u, v)$
"""Return the edge from u to v, or None if not adjacent.""
return self._outgoing[u].get(v) \# returns None if v not adjacent
def degree(self, v, outgoing=True):
""" Return number of (outgoing) edges incident to vertex $v$ in the graph.
If graph is directed, optional parameter used to count incoming edges.
"""
$\operatorname{adj}=$ self._outgoing if outgoing else self._incoming
return len(adj[v])
def incident_edges(self, $v$, outgoing=True):
"""Return all (outgoing) edges incident to vertex $v$ in the graph.
If graph is directed, optional parameter used to request incoming edges.
"""
adj $=$ self._outgoing if outgoing else self._incoming
for edge in adj[v].values( ):
yield edge
def insert_vertex(self, $x=$ None):
"""Insert and return a new Vertex with element $x$."""
v = self.Vertex ( x )
self._outgoing[v] $=\{ \}$
if self.is_directed( ):
self._incoming $[\mathrm{v}]=\{ \} \quad$ \# need distinct map for incoming edges
return $v$
def insert_edge(self, $u, v, x=$ None):
"""Insert and return a new Edge from $u$ to $v$ with auxiliary element $x . "$ ""
$e=$ self.Edge( $u, v, x$ )
self._outgoing $[u][v]=e$
self._incoming $[v][u]=e$

## Thank you.


[^0]:    Figure 2 Co-authorship graph of NiMCS and related research. Nodes represent authors; edges represent co-authorship. Graph layout uses the ForceAtlas2 algorithm. Clusters are calculated via Louvain modularity and delineated by color. Frequency of co-authorship is calculated via

